



Numerical solution of the two-dimensional Poincaré equation

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Received 4 March 2005

Abstract

This paper deals with numerical approximation of the two-dimensional Poincaré equation that arises as a model for internal wave motion in enclosed containers. Inspired by the hyperbolicity of the equation we propose a discretisation particularly suited for this problem, which results in matrices whose size varies linearly with the number of grid points along the coordinate axes. Exact solutions are obtained, defined on a perturbed boundary. Furthermore, the problem is seen to be ill-posed and there is need for a regularisation scheme, which we base on a minimal-energy approach.

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MSC: 65N22; 65F22; 67B55

Keywords: Poincaré equation; Regularisation; Ill-posed problems; Internal waves

1. Introduction

This paper deals with efficient numerical computation of internal wave phenomena in two-dimensional enclosed domains. In contrast to the usual surface waves, internal waves are a type of wave that propagate through a fluid volume. The mechanisms enabling these waves are either rotation or stratification. Efficient numerical models would be of use in various areas of (geophysical) fluid mechanics. One may think of the rotating core of the Earth, or the oceans with their density stratification.

The governing equation can be conveniently expressed in terms of the stream function $\tilde{\Psi}$ (see [17])

$$\frac{\partial^2 \tilde{\Psi}(x, z)}{\partial x^2} - \lambda^2 \frac{\partial^2 \tilde{\Psi}(x, z)}{\partial z^2} = 0 \quad \text{in } \tilde{\Omega},$$

$$\tilde{\Psi}(x, z) = 0 \quad \text{at } \partial\tilde{\Omega}. \quad (1)$$

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In this equation, $\lambda \in \mathbb{R}$ is a parameter corresponding to the frequency at which the waves are forced. The stream function is defined by

$$(u(x, z), v(x, z)) \equiv \left(-\frac{\partial \tilde{\Psi}(x, z)}{\partial z}, \frac{\partial \tilde{\Psi}(x, z)}{\partial x} \right),$$

where $(u(x, z), v(x, z))$ is the velocity vector at a point (x, z) . The fluid flows along streamlines, curves of constant magnitude of the stream function. The boundary $\partial\Omega$ is a simple closed curve in the (x, z) plane. As customary in fluid dynamics, the z -axis points upwards.

In a three-dimensional setting one may derive the equation $\Psi_{xx} + \Psi_{yy} - \lambda^2 \Psi_{zz} = 0$, this equation is commonly called the Poincaré equation, after Cartan [3] who recognised Poincaré as the author who first described the equation in [19]. Although (1) is a reduction of the original three-dimensional problem, we will still refer to it as the Poincaré equation. Some authors prefer the term ‘wave equation’, but we like to stress the absence of a time-like coordinate. This dramatically changes the nature of the problem, instead of an initial value problem we have a boundary value problem.

In this paper, we analyse the solvability of (1) in Sections 2 and 4. Section 3 extends the problem to non-zero values on the boundary. In Section 5, we develop an efficient discretisation scheme. We continue in Section 6 where we will show that the differential equation plus boundary conditions constitute an *ill-posed* problem. We propose a regularisation procedure for the discretised problem, based on minimisation of the energy (Section 7), to obtain meaningful solutions. We conclude by presenting some results of the numerical approximation and regularisation, applied to a representative model geometry.

Previous work on numerical determination of internal waves include [4,12] who use finite differences, respectively, finite element methods. Later attempts use the characteristics of the differential equation, see for example [5–7]. Recent works are that of [21,22], who examine the special case of the spherical shell and [17] where a ray-tracing approach is taken. The existence of *wave attractors* and ill-posedness of the problem is recognised.

Our approach is different, we allow a large class of polynomial domains. Furthermore, we minimise an energy, which acts as a regularisation technique that deals with the ill-posedness of the problem. Previously, authors relied on the inclusion of viscosity as a regularisation technique. The algorithmical approach of [17] does not use regularisation to obtain smooth functions, but instead it gives a recipe for obtaining function values at selected points. Also, this method requires the determination of fundamental intervals (see Section 2), for which there is no general method available. We avoid the use of fundamental intervals. Finally, we believe that in a three-dimensional setting the use of a regularisation method is of even greater importance. We expect the problem to be ill-posed to a higher degree, i.e., more sensitive to disturbances in the relevant parameters, since the behaviour of characteristics is more complex. Also, in principle, our fundamental interval free discretisation method can be generalised to three dimensions. We no longer have a convenient separation of variables, but the principle of discretisation on the boundary only, by considering the mapping of characteristics, is still viable.

Another regularisation technique is inclusion of viscosity in the governing equations. In a physical context this viscosity is a given parameter which is usually very small, which calls for a fine mesh. When the discretisation is refined, there will be more solutions that can be represented on the grid. The existence of many nearby solutions (nearby in parameter space) will still render the problem ill-posed, perhaps not formally, but certainly from a numerical point of view. These issues will be addressed in the forthcoming paper [23]. Also, the delicate dependence of the solution on the shape of the boundary, even in idealised domains, calls for a regularisation technique that is more sophisticated than a fixed viscosity. An energy minimising regularisation procedure like we suggest may perform that function.

There is an extended version of this paper, [24], which includes additional technical calculations that were too lengthy to be included here.

2. The Poincaré equation

In this section we will analyse the Poincaré problem and summarise some previous work. Most importantly, the nature of the solution is seen to be highly dependent on the shape of boundary. There may be no solution at all, or an infinite number of solutions. Furthermore, almost every solution exhibits a fractal structure.

Let us first simplify Eq. (1) by recognising the hyperbolic nature and introducing characteristic coordinates (ξ, η) defined by

$$(\xi(x, z), \eta(x, z)) = \Xi(x, z) \equiv (x - \lambda^{-1}z, x + \lambda^{-1}z).$$

This transforms the domain into $\Omega = \Xi\tilde{\Omega}$ and it transforms (1) to

$$\begin{aligned} \frac{\partial^2 \Psi(\xi, \eta)}{\partial \xi \partial \eta} &= 0 \quad \text{on } \Omega, \\ \Psi(\xi, \eta) &= 0 \quad \text{at } \partial\Omega. \end{aligned} \quad (2)$$

The solvability of (1) is different from (2), there are differing smoothness requirements on solutions. We will, however, be concerned with the latter equation and will not address the issue whether solutions of (2) are also solutions to (1). Let us now define a class of domains.

Definition 1 (*Characteristically convex domains*). We call a domain Ω *characteristically convex*, or convex with respect to the characteristics, if every line in the ξ or in the η direction intersects the boundary $\partial\Omega$ in at most two points.

Note that convexity in the usual sense can be expressed by asking that every line intersects the boundary in at most two points. In this light characteristic convexity is a weaker constraint, every convex domain is characteristically convex. We can now state the following theorem.

Theorem 1. Every solution of the Poincaré equation (2) on a characteristically convex domain Ω is of the form $\Psi(\xi, \eta) = \mathcal{F}(\xi) + \mathcal{G}(\eta)$.

Proof. Choose the origin somewhere in Ω and integrate

$$\int_0^\xi \int_0^\eta \frac{\partial^2 \Psi(r, s)}{\partial r \partial s} dr ds = 0.$$

This yields

$$\Psi(\xi, \eta) - \Psi(\xi, 0) - \Psi(0, \eta) + \Psi(0, 0) = \tilde{\mathcal{F}}(\xi) + \tilde{\mathcal{G}}(\eta),$$

for arbitrary $\tilde{\mathcal{G}}$ and $\tilde{\mathcal{F}}$. The integration can be carried out, since by characteristic convexity the rectangle $(0, \xi) \times (0, \eta)$ is in Ω if the origin and (ξ, η) are in Ω . If $\tilde{\mathcal{F}}(\xi)$ and $\tilde{\mathcal{G}}(\eta)$ are arbitrary then also $\mathcal{F}(\xi) = \tilde{\mathcal{F}}(\xi) + \Psi(\xi, 0) - \Psi(0, 0)$ and $\mathcal{G}(\eta) = \tilde{\mathcal{G}}(\eta) + \Psi(0, \eta) - \Psi(0, 0)$ are arbitrary and we have

$$\Psi(\xi, \eta) = \mathcal{F}(\xi) + \mathcal{G}(\eta). \quad \square$$

Let $c(t) : [0, L] \rightarrow \partial\Omega$ be a piecewise C^k , globally continuous, parametrisation of the boundary,

$$\partial\Omega = \{(\xi(t), \eta(t)) | (\xi(t), \eta(t)) = c(t), 0 \leq t < L, c(0) = c(L)\},$$

where L is the total Euclidean length of $\partial\Omega$. We consider domains with the following properties:

- We only allow simply connected, domains Ω , with closed boundary $\partial\Omega$ as described above, that are convex with respect to the characteristics.
- Segments of the boundary are not allowed to be parallel to the coordinate axes, i.e., the sets $\{t | c'(t)^T e_1 = 0\}$ and $\{t | c'(t)^T e_2 = 0\}$ have measure zero within $\partial\Omega$.

Next we define the vertices of the domain,

Definition 2 (*Corners*). If $\lim_{t \uparrow a} c'(t) \neq \lim_{t \downarrow a} c'(t)$ we call the point $c(a)$ a *corner point* or *vertex*. Define the *characteristic rectangle* $R = (\xi^-, \xi^+) \times (\eta^-, \eta^+)$ as the smallest rectangle containing Ω . We call the four points $\bar{\Omega} \cap \bar{R}$ the *extreme vertices*.

The solution to the Poincaré equation can be extended to the characteristic rectangle.

Theorem 2 (Extension of the solution). *Every solution to the Poincaré equation on $\bar{\Omega} = \Omega \cup \partial\Omega$ is extendible to $\bar{R} = [\xi^-, \xi^+] \times [\eta^-, \eta^+]$.*

Proof. For every point $(\xi, \eta) \in \bar{R} \setminus \bar{\Omega}$ it is possible to find ξ' and η' such that $(\xi', \eta) \in \bar{\Omega}$ and $(\xi, \eta') \in \bar{\Omega}$. These points are inside the domain and we have $\Psi(\xi', \eta) = \mathcal{F}(\xi') + \mathcal{G}(\eta)$ and $\Psi(\xi, \eta') = \mathcal{F}(\xi) + \mathcal{G}(\eta')$. Thus, $\mathcal{F}(\xi)$ and $\mathcal{G}(\eta)$ exist and can be used to define $\Psi(\xi, \eta) = \mathcal{F}(\xi) + \mathcal{G}(\eta)$. \square

We have one extra constraint from the physics of the internal wave problem. The boundary condition $\Psi = 0$ comes from the fact that there should be no flow through the boundary, expressed by $(u, v) \cdot n = 0$ where n is the outward unit normal to the boundary. At a corner point the normal is not defined, but physically the no-flow condition dictates that $(u, v) = 0$. We will add this as an extra constraint. In terms of the stream function this becomes

$$\nabla \Psi = 0 \quad \text{at a corner point.} \quad (3)$$

Since we extended the domain to \bar{R} the gradient operator is well defined except for the at most four extreme vertices where we may use one-sided derivatives for \mathcal{F} or \mathcal{G} .

We know that the stream function Ψ may be scaled, multiplicatively, by any constant, thereby scaling the velocities. We do not care about the difference between Ψ and $c\Psi$, and we normalise the stream function as

$$\|\Psi\|_{L_2(\Omega)}^2 = \int_{\Omega} \Psi^2(\xi, \eta) d\xi d\eta = \langle \mathcal{F} + \mathcal{G}, \mathcal{F} + \mathcal{G} \rangle_{L_2(\Omega)} = 1. \quad (4)$$

This constraint also rules out the trivial solution $\Psi = 0$, and by boundary conditions the constant solution $\Psi = c$.

Furthermore, we want a unique representation of Ψ in terms of \mathcal{F} and \mathcal{G} . Suppose there is another pair $\tilde{\mathcal{F}}, \tilde{\mathcal{G}}$, for which

$$\mathcal{F} + \mathcal{G} = \tilde{\mathcal{F}} + \tilde{\mathcal{G}} \Rightarrow \mathcal{F} - \tilde{\mathcal{F}} = \tilde{\mathcal{G}} - \mathcal{G}.$$

Since ξ and η may be independently varied, we have that $\mathcal{F} = \tilde{\mathcal{F}} + c$ and $\mathcal{G} = \tilde{\mathcal{G}} - c$ for an arbitrary constant c . In order to have a unique constant we normalise

$$\langle \mathcal{F}, 1 \rangle_{L_2([\xi^-, \xi^+])} = 0. \quad (5)$$

The Poincaré problem is now reduced to finding \mathcal{F} and \mathcal{G} for which

$$\begin{aligned} \mathcal{F}(\xi) + \mathcal{G}(\eta) &= 0 \quad \text{at } \partial\Omega, \\ \mathcal{F}'(\xi) &= \mathcal{G}'(\eta) = 0 \quad \text{at a corner point,} \\ \langle \mathcal{F}, 1 \rangle_{L_2(\Omega)} &= 0, \\ \langle \mathcal{F} + \mathcal{G}, \mathcal{F} + \mathcal{G} \rangle_{L_2(\Omega)} &= 1, \end{aligned} \quad (6)$$

with $\mathcal{F} \in C^1 : [\xi^-, \xi^+] \rightarrow \mathbb{R}$ and $\mathcal{G} \in C^1 : [\eta^-, \eta^+] \rightarrow \mathbb{R}$.

We will show that the first requirement in (6) induces a map from the boundary to itself. This was firstly noted in [13] and later constructively applied in [17]. This can be observed by considering function values $\mathcal{G}(\eta_1) = -\mathcal{F}(\xi_1)$ at a boundary point (ξ_1, η_1) . When tracing a line in the ξ -direction a new boundary intersection (ξ_2, η_2) is found. Of course $\eta_2 = \eta_1$ and we find $\mathcal{F}(\xi_2) = -\mathcal{G}(\eta_2) = -\mathcal{G}(\eta_1) = \mathcal{F}(\xi_1)$. Carrying through with this procedure we find that one function value on the boundary determines function values at boundary points $(\xi_1, \eta_1), \dots, (\xi_n, \eta_n)$. Much work has been done in studying the Poincaré problem by viewing the above construct as a dynamical system. We will review the main results on existence of solutions in the case that the boundary satisfies the assumption of characteristic convexity. In this case we can follow [13] and introduce the following homeomorphisms:

- T^+ assigns to a boundary point the unique boundary point with the same η -coordinate, with the exception that the top and bottom extremal vertices are mapped onto themselves.
- T^- assigns to a boundary point the unique boundary point with the same ξ -coordinate, with the exception that the left and right extremal vertices are mapped onto themselves.
- $F = T^- \circ T^+ : \partial\Omega \rightarrow \partial\Omega$.

Note that F is an orientation preserving map. We now define an orbit as

$$O(P) \equiv \{P, T^+P, FP, (T^+ \circ F)P, F^2P, \dots\} \quad \text{where } P \in \partial\Omega.$$

Let $S(P) \equiv c^{-1}(P)|_{[0,L]}$ be the coordinate (by arc length distance) of a point P on $\partial\Omega$. By characteristic convexity of the boundary, we can lift the homeomorphism F to a continuous increasing function $f: \mathbb{R} \rightarrow \mathbb{R}$ such that

$$f(s+L) = f(s) + L, \quad s \in \mathbb{R}$$

and

$$S(FP) = f(S(P)) \pmod{L}, \quad P \in \partial\Omega.$$

If we now set $f_k(s) = f(f_{k-1}(s))$ where $f_1(s) = f(s)$ we can define the *rotation number* $\alpha(F)$ of F by

$$\alpha(F) \equiv \lim_{n \rightarrow \infty} \frac{f_n(s)}{n} \in [0, L).$$

The rotation number can be shown to exist and to be independent of s [8]. A classical theorem due to [13] is

Theorem 3. *The homeomorphism F has three separate types of behaviour in a characteristically convex domain*

- (A) *We have $\alpha(F) = m/n$ for some $m, n \in \mathbb{N}$ and $F^n = I$. This case is often called a resonance, or modal solution.*
- (B) *We have $\alpha(F) = m/n$ for some $m, n \in \mathbb{N}$, but $F^n \neq I$. The dynamical system defined by F has a finite number of attracting fixed points. We call solutions of the Poincaré equation in this case wave attractors.*
- (C) *$\alpha(F) \in \mathbb{R} \setminus \mathbb{Q}$, then F^k has no fixed point for any k . There is only one orbit that fills the entire boundary. This is the ergodic case.*

Note that for a given $\alpha(F)$ the values of n and m are not unique in the first two cases. Multiples km and kn of m and n will yield the same rotation number and will not change the solution type. The solvability of the problem (2) depends on the case in which the rotation number falls. This delicate dependence on the rotation number, and thus on the geometry of the domain, makes the problem ill-posed. This will be of concern when considering a numerical approximation technique. To proceed, we need the concept of *fundamental interval* [17,14]. The fundamental interval can be defined as the set $M \subset \partial\Omega$ for which every orbit has at most one point in M and all orbits that have a point in M constitute the entire boundary. The following definition in [16] gives a more constructive approach:

Definition 3 (Fundamental interval). Denote by $(P, Q)_{\partial\Omega}$ the open set of points in between P and Q (using anti clockwise orientation) on the boundary. Set $(P, Q]_{\partial\Omega} \equiv (P, Q)_{\partial\Omega} \cup Q$, $[P, Q)_{\partial\Omega} \equiv (P, Q)_{\partial\Omega} \cup P$ and $[P, Q]_{\partial\Omega} \equiv (P, Q)_{\partial\Omega} \cup Q \cup P$. Denote the vertices by P_0, \dots, P_3 , with P_0 the uppermost extreme vertex and P_1, \dots, P_3 the rest of the vertices in anti-clockwise order. Now define, for the same n as used in (3), in the case (A), the following

- for n even: P^* as the point from the finite set $O(P_1) \cap (P_0, P_1]_{\partial\Omega}$ such that $O(P_1) \cap (P_0, P^*)_{\partial\Omega} = \emptyset$,
- for n odd: P^* as the point from the finite set $O(P_2) \cap (P_0, P_1]_{\partial\Omega}$ such that $O(P_2) \cap (P_0, P^*)_{\partial\Omega} = \emptyset$.

The fundamental interval¹ is then given by $M = [P_0, P^*)$. Define M_ξ and M_η by orthogonal projection of M on the ξ and η axes.

In order to clarify the above definition an example has been constructed in Fig. 1. It was proved for the resonant case that the fundamental interval must be supplied with boundary conditions (for either \mathcal{F} or \mathcal{G}) in order to obtain a unique solution. It is likely that the concept of the fundamental interval is also useful in the case (B), the attractor regime. In [17] a procedure is suggested for several geometries. In the following, we will work with the assumption that the fundamental intervals also exist in the wave attractor case.

When in the following discussion we speak of Dirichlet boundary conditions, we mean choosing a function $h(\xi)$ and setting $\mathcal{F}(\xi) = h(\xi)$ on M_ξ . This implies the value for \mathcal{G} on M_η by the boundary condition $\mathcal{F}(\xi) + \mathcal{G}(\eta) = 0$. The

¹ In [16] the fundamental interval is called *generating set*.

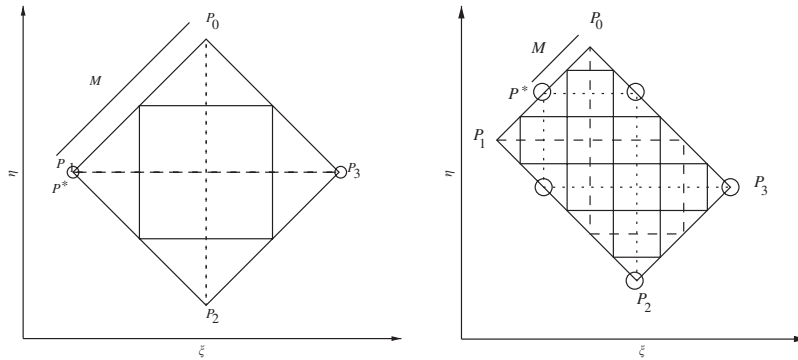


Fig. 1. This figure clarifies the fundamental interval as introduced in Definition 3. At the right is shown a geometry with $n = 5$, which is easy to see by following the closed orbit (solid line). Of course, one may also consider the dashed and dotted lines that connect $O(P_1) = O(P_0)$, respectively, $O(P_2) = O(P_3)$. Keep in mind that $T^+P_0 = P_0$, $T^-P_1 = P_1$, etc. The orbit $O(P_2)$ is indicated by circles. The left picture is a case where $n = 2$. In both pictures, the set $M = [P_0, P^*)$ is the fundamental interval.

solvability of (2), together with this additional boundary condition, is different for the three cases in Theorem 3. Some results are summarised in the following theorem.

Theorem 4 (Solvability in L_2). *The solvability of the Poincaré problem (2) in $L_2(\Omega)$ is dependent on the cases, and the value of n , from Theorem 3 as follows:*

- (A) *If Dirichlet data $h(\xi)$ for $\mathcal{F}(\xi)$ are supplied on M_ξ , then we have an unique solution [14, Theorem 6]. In this case, there exist piecewise C^k coordinate transforms $\xi \rightarrow p(\xi)$ and $\eta \rightarrow q(\eta)$ that transform the domain to a rotated rectangle [15,16]. In (x, z) coordinates this corresponds to a non-rotated rectangle for which the solutions are known. The solutions on the original domain are then found using inverse coordinate transforms.*
- (B) *According to [9, Theorem 2, Remark 2] the problem is solvable.*
- (C) *From [9, Theorem 5, Remark 5] there exists only the trivial solution in this case.*

The conclusions in this theorem can be extended to the situation where an inhomogeneous term $f(\xi, \eta, \Psi)$ is added to (2). The conclusions do have to be modified, for example, solutions are possible in the case (C) if there exists a $C(\alpha) > 0$ such that $|\alpha - m/n| \geq C(\alpha)/n^2$ for any rational number m/n , where α is the rotation number [9].

3. Boundary forcing

The Poincaré problem as posed before describes *free vibrations*, i.e., eigenmodes of the system. In this case, however, the term ‘eigenmode’ may be inappropriate since the spectrum often has continuous parts, or the spectrum might be dense in an interval. The spectral properties of the Poincaré problem are more extensively investigated in [20].

We present two approaches for selecting relevant solutions from the infinitude of possibilities. First we present an energy minimizing regularisation procedure for the Poincaré equation

$$\begin{aligned} \Psi_{xx} - \lambda^2 \Psi_{zz} &= 0 \quad \text{in } \Omega, \\ \Psi &= 0 \quad \text{at } \partial\Omega. \end{aligned} \tag{7}$$

This leads to a minimisation problem of the form

$$x_\tau = \arg \min (\|Ax\|_2^2 + \tau^2 \|Lx\|_2^2),$$

where A discretises the Poincaré problem and L measures the energy of the system. Sections 5 and 6 deal with this regularisation problem.

One may also hold the view that the eigenmodes that occur in nature are determined by some kind of forcing. The simplest case is a forcing operating on the boundary of the domain only. Examples include tidal forcing on ocean surfaces or some mechanical forcing. The problem to be solved is in this case

$$\begin{aligned}\Psi_{xx} - \lambda^2 \Psi_{zz} &= 0 \quad \text{in } \Omega, \\ \Psi &= h(\xi, \eta) \quad \text{at } \partial\Omega,\end{aligned}\tag{8}$$

leading to a minimisation problem of the form

$$x_\tau = \arg \min(\|Ax - b\|_2^2 + \tau^2 \|Lx\|_2^2),\tag{9}$$

where b is the discrete discretised right-hand side h . Note that any solution of (7) may be arbitrarily added to a solution of the forced problem. In Section 2, it was shown that the characteristics lead to functional relationships between \mathcal{F} and \mathcal{G} . The forced problem similarly yields functional dependencies, involving the function h . We add a bar to solutions of the forced problem from here on. In terms of $\bar{\mathcal{F}}$ and $\bar{\mathcal{G}}$ the problem reads

$$\bar{\mathcal{F}} + \bar{\mathcal{G}} = h(\xi, \eta) \quad \text{at } \partial\Omega.\tag{10}$$

The restrictions on \mathcal{G} and $\bar{\mathcal{G}}$ induced by the characteristic connecting (ξ_1, η_1) with (ξ_1, η_2) look like

$$\begin{aligned}\bar{\mathcal{G}}(\eta_1) - \bar{\mathcal{G}}(\eta_2) &= h(\xi_1, \eta_1) - h(\xi_1, \eta_2) \quad \text{for forced problem,} \\ \mathcal{G}(\eta_1) - \mathcal{G}(\eta_2) &= 0 \quad \text{for the unforced problem,}\end{aligned}\tag{11}$$

and we see that indeed $\mathcal{G} + \bar{\mathcal{G}}$ solves the forced problem. The only open question is the role of the fundamental interval and the manner in which the fractal nature of solutions presents itself to us in the forced setting. In the homogeneous problem we would, for example, conclude from (11) that $\mathcal{G}(\eta_1)$ must be equal to $\mathcal{G}(\eta_2)$. Further analysis, as sketched above, revealed that there exist fundamental intervals where it is necessary and sufficient to prescribe \mathcal{G} (or \mathcal{F}). Any homogeneous solution that we add to the forced problem will have fundamental intervals and associated fractal structure. Working out the functional relations for the forced problem reveals that

$$\begin{aligned}\bar{\mathcal{F}}_i + \bar{\mathcal{G}}_i &= h_{2i}, \\ \bar{\mathcal{F}}_i + \bar{\mathcal{G}}_{i+1} &= h_{2i+1},\end{aligned}\tag{12}$$

where $\mathcal{F}_i = \mathcal{F}(\xi_i)$, $\mathcal{G}_i = \mathcal{G}(\eta_i)$ and the h_i are the function h , evaluated at the points visited while tracing the characteristics. For \mathcal{F}_i and \mathcal{G}_i this implies

$$\begin{aligned}\bar{\mathcal{G}}_{i+1} &= \bar{\mathcal{G}}_i + h_{2i+1} - h_{2i}, \\ \bar{\mathcal{F}}_{i+1} &= \bar{\mathcal{F}}_i + h_{2i+2} - h_{2i+1}\end{aligned}$$

for the relation between the function values separated by one step. Suppose we know $\bar{\mathcal{F}}_0, \bar{\mathcal{G}}_0$, we may then solve the recursion and obtain

$$\begin{aligned}\bar{\mathcal{G}}_i &= \bar{\mathcal{G}}_0 + \sum_{j=0}^{2i-1} (-1)^{j+1} h_j, \\ \bar{\mathcal{F}}_i &= \bar{\mathcal{F}}_0 + \sum_{j=1}^{2i} (-1)^j h_j,\end{aligned}$$

which tells us that boundary values are alternately added and subtracted from the initial values \mathcal{G}_0 and \mathcal{F}_0 . This behaviour was noted before in [1]. Note that by adding together the expressions we recover (12), as required. If the prescribed function h is non-zero in the neighbourhood of the attractor, then the limit for $i \rightarrow \infty$ is likely to be non-existent. Just like in the unforced case we have fundamental intervals where values of $\bar{\mathcal{F}}$ or $\bar{\mathcal{G}}$ may be arbitrarily prescribed. The forced problem is still under determined and sensitive to the shape of the domain, thus it is still an ill-posed problem.

4. Structure of solutions

Except for the classification of solutions into attractors and resonances we can get further information by viewing the Poincaré equation as a Hamiltonian system. Choose any closed curve $(\xi(t), \eta(t)) \in R$. Denote the derivative with respect to time by a dot. Take the stream function as the Hamiltonian, then

$$\dot{\Psi}(\xi, \eta) = \mathcal{F}'(\xi)\dot{\xi} + \mathcal{G}'(\eta)\dot{\eta}, \quad (13)$$

At curves $\{(\xi(t), \eta(t)) | \dot{\Psi} = 0\}$ the stream function Ψ is constant. Since $\partial\Psi/\partial\xi = \mathcal{F}'$ and $\partial\Psi/\partial\eta = \mathcal{G}'$ we see from (13) that this is the case when

$$\begin{pmatrix} \dot{\xi} \\ \dot{\eta} \end{pmatrix} = \begin{pmatrix} \mathcal{G}'(\eta) \\ -\mathcal{F}'(\xi) \end{pmatrix}. \quad (14)$$

This system is in Hamiltonian form. We proceed by linearising \mathcal{F} and \mathcal{G} around $(\xi, \eta) = (0, 0)$, giving

$$\mathcal{F}'(\xi) = \mathcal{F}'(0) + \xi\mathcal{F}''(0) + \mathcal{O}(\xi^2),$$

$$\mathcal{G}'(\eta) = \mathcal{G}'(0) + \eta\mathcal{G}''(0) + \mathcal{O}(\eta^2).$$

We suppose that \mathcal{F} and \mathcal{G} are non-degenerate at the origin, i.e., $\mathcal{F}(0)'' \neq 0$ and $\mathcal{G}(0)'' \neq 0$. The system (14) is now transformed to

$$\begin{pmatrix} \dot{\xi} \\ \dot{\eta} \end{pmatrix} = \begin{pmatrix} 0 & \mathcal{G}''(0) \\ -\mathcal{F}''(0) & 0 \end{pmatrix} \begin{pmatrix} \xi \\ \eta \end{pmatrix} + \begin{pmatrix} \mathcal{G}'(0) \\ -\mathcal{F}'(0) \end{pmatrix} + \begin{pmatrix} \mathcal{O}(\eta^2) \\ \mathcal{O}(\xi^2) \end{pmatrix}.$$

The coordinates may be shifted, $(x, y) = (\xi + a, \eta + b)$ to make the problem homogeneous. The choices $a = -\mathcal{F}'(0)/\mathcal{F}''(0)$ and $b = \mathcal{G}'(0)/\mathcal{G}''(0)$ lead to

$$\begin{pmatrix} \dot{x} \\ \dot{y} \end{pmatrix} = A \begin{pmatrix} x \\ y \end{pmatrix} + \begin{pmatrix} \mathcal{O}(y^2) \\ \mathcal{O}(x^2) \end{pmatrix} \quad \text{where } A = \begin{pmatrix} 0 & \mathcal{G}''(0) \\ -\mathcal{F}''(0) & 0 \end{pmatrix}.$$

The eigenvalues of A are $\lambda_{1,2} = \pm\sqrt{-\mathcal{F}''(0)\mathcal{G}''(0)}$ which makes the level set of $\Psi(x, y)$ either of saddle or of centre type. We call points where $\nabla\Psi = 0$ critical points, this happens, for example, at corner points. At these points we have potentially a local extremum for Ψ and the shift is $a = b = 0$.

However, if the origin is not critical, but degenerate ($\mathcal{F}''(0) = \mathcal{G}''(0) = 0$), then it is easy to see that

$$\xi = \mathcal{G}'(0)t + t\mathcal{O}(\eta^2) + c_1,$$

$$\eta = -\mathcal{F}'(0)t + t\mathcal{O}(\xi^2) + c_2,$$

for arbitrary constants c_1 and c_2 . Close to the chosen origin the level sets of Ψ are approximately straight lines. If we have

$$\mathcal{F}'(0) = \mathcal{G}'(0) = 0 \quad \text{and} \quad \mathcal{F}''(0) = 0 \quad \text{or} \quad \mathcal{G}''(0) = 0, \quad (15)$$

then we need to look at higher-order expansions. We will not consider this situation here. We summarise the results:

- Suppose $\Psi(0, 0) = c$ and $\mathcal{F}''\mathcal{G}'' < 0$, then the level set of $\Psi = c$ is a hyperbola close to the critical point where $\nabla\Psi = 0$, this is a saddle point.
- Suppose $\Psi(0, 0) = c$ and $\mathcal{F}''\mathcal{G}'' = 0$, then the level set of $\Psi = c$ is a curve with slope $\mathcal{F}'(0)/\mathcal{G}'(0)$ at the origin which is not a critical point.
- Suppose $\Psi(0, 0) = c$ and $\mathcal{F}''\mathcal{G}'' > 0$, then the level set of $\Psi = c$ is an ellipse around the extremum where $\nabla\Psi = 0$. This extremum can only occur in the interior of the domain.

All results are valid for $(\xi, \eta) \in R$, which excludes the extreme vertices. The invariance of \mathcal{G} and \mathcal{F} along, respectively, the ξ and η coordinate axes leads to the following theorems.

Lemma 1. *If we are not in the case (15), then every corner point is a saddle point, including the extreme vertices if they are corner points.*

Proof. Every corner point is a critical point, since $\nabla\Psi = 0$ by (3). Take one-sided derivatives at the extreme vertices. Centre points are excluded since we are on the boundary. Since we are not in the case (15) the only option is a saddle point. \square

Now, let $l(P)$ be a tangent vector to the boundary at a point $P \in \partial\Omega$, let l be multi-valued at corner points.

Theorem 5. *If a point P at the boundary is a saddle point, then the points in the sequences (T^+P, T^-T^+P, \dots) and (T^-P, T^+T^-P, \dots) are also saddles. If a sequence has a finite number of distinct members, then the final point Q is such that $l^T(T^-Q)e_2 = 0$ or $l^T(T^+Q)e_1 = 0$.*

Proof. If P is a saddle then $\mathcal{F}' = 0$ and $\mathcal{F}'' \neq 0$. At the boundary point $D = T^-(P)$ this also holds. Furthermore, at this point $\nabla\Psi(D) \cdot l(P) = 0$. If the η component of $l(D)$ is not zero, it must be the case that $\mathcal{G}' = 0$. We thus have $\nabla\Psi = 0$, the point is a critical point. Since centre points cannot exist on the boundary the point must be a saddle. If the η component of $l(D)$ were zero the procedure stops. If not, then we continue with the boundary point with the same η coordinate as D . \square

Theorem 6. *The points (ξ_1, η_1) and (ξ_2, η_2) in R are critical points if and only if (ξ_1, η_2) and (ξ_2, η_1) are critical points in R .*

Proof. Again, use the invariance of \mathcal{F} and \mathcal{G} in the η and ξ directions to find that $\nabla\Psi = 0$ at (ξ_1, η_2) and (ξ_2, η_1) and that $\mathcal{F}''\mathcal{G}'' \neq 0$ at these points. \square

How these results help us is perhaps best demonstrated by some examples.

Example 1 (The circle $\xi^2 + \eta^2 = 1$). The simplest solution would be one centre in the middle. The extremal points with the same ξ and η coordinates have $l^Te_1 = 0$ or $l^Te_2 = 0$ and thus need not be saddle points. If we allow two critical points at the boundary, they must be at opposite extremal points or we have four critical points by Theorem 6. Of each saddle, one of the branches is part of the boundary, while the other branch crosses the boundary. Connecting each saddle to itself yields a critical point on the boundary that makes the boundary non-smooth. Therefore, the two zero level set curves that point into the circle must be connected. The circle is now divided in two cells, a centre point must be inside each cell. One may continue in this fashion to obtain more possibilities. These solutions correspond topologically to the solutions given in [2].

Example 2 (The rotated rectangle). Consider a rectangle of width a and length $2a$, rotated over 45° . By Theorem 5 the saddles at the corner points induce saddles at the midpoints of the long sides of the rectangle. By Theorem 6 two critical points must exist inside the domain. They can be centres, giving rise to a pattern with two cells. They cannot be saddles, the stable and unstable manifold would have no option but forming homoclinic connections. Inside the loop there have to be centres which in turn induce saddles on the boundary. This cannot be done without violating the flow directions.

By placing extra saddles on the boundary and centres in the domain one obtains $j \times 2j$ cells, in accordance with the theory [17].

5. Discretisation

In this section we will develop a discretisation for Eq. (10). The discretisation will be posed in the (ξ, η) -coordinate frame, since here the Poincaré equation becomes separable. Approximations $\tilde{\mathcal{F}}$ and $\tilde{\mathcal{G}}$ of \mathcal{F} and \mathcal{G} will be sought in the spaces $V_\xi \subset H^1([\xi^-, \xi^+])$ and $V_\eta \subset H^1([\eta^-, \eta^+])$, spanned by piecewise polynomial basis functions $\phi_i(\xi)$,

respectively, $\psi_j(\eta)$. In this basis we can write

$$\begin{aligned}\tilde{\mathcal{F}}(\xi) &= \sum_{i=1}^n f_i \phi_i(\xi) \in V_\xi, \\ \tilde{\mathcal{G}}(\eta) &= \sum_{j=1}^m g_j \psi_j(\eta) \in V_\eta.\end{aligned}\tag{16}$$

We now define our approximation $\tilde{\Psi}$ to Ψ by

$$\tilde{\Psi}(\xi, \eta) = \tilde{\mathcal{F}}(\xi) + \tilde{\mathcal{G}}(\eta).\tag{17}$$

We propose a Galerkin orthogonality condition of the system (10) and try to find $\tilde{\mathcal{F}}$ and $\tilde{\mathcal{G}}$ such that the residual $H = \mathcal{F} + \mathcal{G} - h$ is orthogonal to all test functions $v_1 \in V_\xi$ and $v_2 \in V_\eta$ on the boundary $\partial\Omega$. First, define on $\partial\Omega$ the line integral

$$(u, v) = \int_{\partial\Omega} uv \, dl.$$

The orthogonality relation becomes

$$(H, v) = (\tilde{\mathcal{F}} + \tilde{\mathcal{G}}, v) - (h, v) = 0 \quad \forall v \in V,$$

where $V = \{w(\xi) + v(\eta) | w \in V_\xi, v \in V_\eta\}$. Equivalently, the orthogonality relation can be written as

$$\begin{aligned}(\tilde{\mathcal{F}}, v_1) + (\tilde{\mathcal{G}}, v_1) &= (h, v_1) \quad \forall v_1 \in V_\xi, \\ (\tilde{\mathcal{F}}, v_2) + (\tilde{\mathcal{G}}, v_2) &= (h, v_2) \quad \forall v_2 \in V_\eta.\end{aligned}\tag{18}$$

We have to test against all functions in V_ξ and V_η , we choose the basis functions ϕ_i and ψ_j . This yields

$$\begin{aligned}\sum_{i=1}^n f_i (\phi_i, \phi_j) + \sum_{i=1}^m g_i (\psi_i, \phi_j) &= (h, \phi_j), \quad j = 1, \dots, n, \\ \sum_{i=1}^n f_i (\phi_i, \psi_j) + \sum_{i=1}^m g_i (\psi_i, \psi_j) &= (h, \psi_j), \quad j = 1, \dots, m.\end{aligned}\tag{19}$$

After choosing suitable spaces V_ξ and V_η we can work out the integrations in (19). This results in a matrix vector equation which we need to solve for $f = (f_1, \dots, f_n)^T$ and $g = (g_1, \dots, g_m)^T$:

$$\begin{pmatrix} A_1 & A_2 \\ A_2^T & A_3 \end{pmatrix} \begin{pmatrix} f \\ g \end{pmatrix} = A \begin{pmatrix} f \\ g \end{pmatrix} = \begin{pmatrix} h_1 \\ h_2 \end{pmatrix}.\tag{20}$$

After discretisation, the dimension of the matrix A will be much smaller than the dimension that would have been obtained by using, for example, standard finite element or finite difference methods on (6). For a resolution of $m \times n$ grid points we will only require matrices of dimension $m + n$, whereas standard methods yield a dimension of order mn . It is nice to see that the non-zero pattern of A_2 mimics the shape of the boundary. This is clear, since if $(A_2)_{ij} \neq 0$ then the intersection of the supports of ϕ_i and ψ_j contain the boundary. The sub-matrices A_1 and A_3 are banded since only the supports of $\psi_{i-N}, \dots, \psi_{i+N}$ overlap the support of ψ_i , with N depending on the specific basis functions used. See Fig. 2 for an example.

The calculation of the matrix A is detailed in [24] for the case of piecewise linear functions and a polygonal domain Ω .

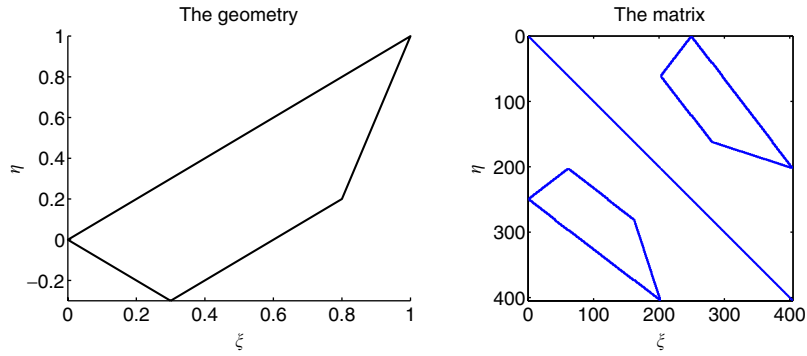


Fig. 2. This figure shows a sample geometry in the left panel. To the right is a plot of the non-zero elements of the matrix A . The geometry stands out clearly in the blocks A_2 and A_2^T (see (20)). The sub-matrices A_1 and A_3 are tri-diagonal since standard piecewise linear basis functions were used.

The constraint $\langle \mathcal{F}, 1 \rangle_{L_2([\xi^-, \xi^+])} = 0$ is easily discretised as

$$\left(\sum_{i=1}^n e_i^T \right) f = 0.$$

We will express this in matrix form as $Cx = 0$, with $C \in \mathbb{R}^{2(n+m)}$ and x the concatenation of the vectors f and g . The constraint $\|\Psi\|_{L_2(\Omega)} = 1$ can be discretised as $\|Nx\|_2^2 = 1$. We do not need N explicitly, since if we have a solution x with $\|Nx\| = 1$, then there is a constant c such that $\|cx\|_2^2 = 1$. The precise value of c is immaterial, as long as the norm of the solution is fixed at some value, therefore, we fix the norm using

$$\|x\|_2^2 = 1. \quad (21)$$

Finally, we need to discretise the constraint at corner points, $\nabla \Psi = 0$. We use finite differences as follows:

$$\tilde{\mathcal{F}}'(\xi_i) = \begin{cases} \frac{f_{i+1} - f_i}{\xi_{i+1} - \xi_i} & \text{if } \xi_i = \xi^-, \\ \frac{f_i - f_{i-1}}{\xi_i - \xi_{i-1}} & \text{otherwise.} \end{cases}$$

The derivative of $\tilde{\mathcal{G}}(\eta)$ is defined accordingly. We write a matrix vector equation

$$Px = 0. \quad (22)$$

Rows of P express that $f_i = f_{i-1}$ (or $f_i = f_{i+1}$ at $\xi_i = \xi^-$) or $g_j = g_{j-1}$ (or the exception $g_j = g_{j+1}$ at $\eta_j = \eta^-$) if (ξ_i, η_j) is a corner point.

Note that solutions to the discretised problem are *exact* solutions to the Poincaré equation, since they are of the form $\mathcal{F}(\xi) + \mathcal{G}(\eta)$. If the residual is non-zero, then the zero level set of this solution will not coincide with the boundary $\partial\Omega$. We obtain an exact solution on a modified boundary. Unfortunately, the perturbation of the zero level set of $\mathcal{F} + \mathcal{G}$ is not necessarily small, even though $\mathcal{F} + \mathcal{G}$ is small on $\partial\Omega$. The boundary might ‘fold open’ if the zero level set was at a saddle point and follows nearby level sets in the discretised case.

6. Ill-Posedness and uniqueness

After discretisation of the Poincaré equation with $\Psi = 0$ at the boundary, the problem reduces to solving $Ax = 0$. The ill-posedness of the Poincaré equation is reflected in the matrix A . It has singular values rapidly decreasing towards zero. There are many singular values close to zero, their number increasing with increasing grid size. To these singular values correspond singular vectors that are close to the null space of A in the sense that the residual $\|Av\|_2^2$ is small

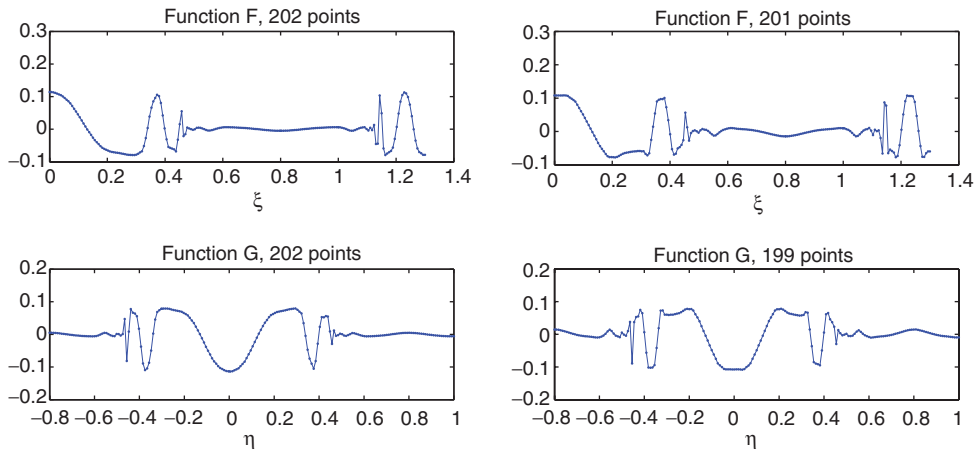


Fig. 3. This figure illustrates that a small change in the discretisation can have substantial effects on the solution. Plotted are the approximations to \mathcal{F} and \mathcal{G} . The panel to the left has 200 grid points in both the ξ and η direction, the right panel has one grid point less in both directions. The two solutions are quite different.

for such a singular vector v . We exclude the trivial solution and we pose the problem as a minimisation of $\|Ax\|_2^2$. There is a large sensitivity of the solution x to small perturbations in the matrix A . This is the discrete analogue of the delicate dependence of the solution to changes in the boundary. We will find solutions close to solutions from the infinite dimensional solution space of the Poincaré equation, yet we cannot control which discrete vectors x we obtain. The specific vectors found are dependent on the discretisation. When the discretisation is slightly changed, for example, by adding some grid points, completely different valid solutions may emerge (see Fig. 3).

Typically, continuous ill-posed problems have no gaps in their spectrum. The discretised problem tends to inherit this property, and a truncation of the singular values in order to decide what does or does not belong to the kernel of A becomes infeasible, since there is no clear cut-off point.

Also, we know that there is a delicate dependence of the solvability on the rotation number of the domain. Slight changes in the parameter λ might change the nature of the solution. We anticipate the possibility of contamination of the numerical null space by unwanted components of solutions corresponding to nearby values of λ .

We use a regularisation method that in some sense relaxes the requirement that $\mathcal{F} + \mathcal{G} = 0$ at the boundary. We find solutions where $\mathcal{F} + \mathcal{G}$ is small at the boundary. Equivalently, we can say that we solve the Poincaré equation on a domain that is slightly perturbed from Ω , the level curve of $\Psi = 0$ is likely to be close to $\partial\Omega$. Instead of asking that $\mathcal{F} + \mathcal{G}$ be zero at boundary we try to minimise this quantity, while at the same time we try to obtain a smooth solution by minimising the energy. The balance between these two goals is tuned using a regularisation parameter. This parameter is picked using a tool called the *L-curve* (see e.g., [11]), which we discuss in Section 6.1. The important point is that a small value of the regularisation parameter indicates that little smoothing was required, the value of λ is probably very close to a resonance. If in contrast the value is rather high, then we must be far away from a resonance. Maybe we are in the attractor case where high energy is induced by the fractal structure, or maybe the chosen value of λ does not yield solutions at all. Visual inspection of the solutions and interpretation of the *L-curve* usually gives information on these issues.

There exists the freedom to specify boundary conditions on the fundamental interval, which has not been incorporated in our method of discretisation and solution. We will rely on our numerical method to fill the fundamental interval. The first reason for this is that fundamental intervals are a property of the specific geometry under consideration and need to be established whenever a new geometry is considered. A second reason is that fundamental intervals are not unique. If M is a fundamental interval, then any $F^k M$ is also a valid fundamental interval. The question arises to which interval Dirichlet boundary conditions should be applied. The interval that contains the largest number of grid points would be a sensible choice, yet it is not clear if the solution is now unique in any numerical sense.

We will assume that the state possessing *minimal kinetic energy* is the physically relevant solution. Additionally the buoyancy energy could be included, we will not do this for ease of presentation. One can think of the energy as an

ordering principle of the solution space. We define the total kinetic energy operator $\bar{T} : C^1(\Omega) \rightarrow \mathbb{R}$ in the (x, y) system by

$$\bar{T}\bar{\Psi} = \int_{\bar{\Omega}} \nabla \bar{\Psi} \cdot \nabla \bar{\Psi} \, dx \, dy,$$

which is, in characteristic coordinates

$$\begin{aligned} T\Psi &= \int_{\Omega} \left[\left(\frac{\partial}{\partial \eta} + \frac{\partial}{\partial \xi} \right)^2 + \lambda^2 \left(\frac{\partial}{\partial \eta} - \frac{\partial}{\partial \xi} \right)^2 \right] \Psi^2 \frac{1}{2\lambda} \, d\xi \, d\eta \\ &= \frac{1}{2\lambda} \int_{\Omega} (1 + \lambda^2) [(\mathcal{F}')^2 + (\mathcal{G}')^2] + (1 - \lambda^2) \mathcal{F}' \mathcal{G}' \, d\xi \, d\eta. \end{aligned}$$

We can eliminate the cross term by using partial integration in the form

$$\int_{\Omega} (\partial_i u) v \, dA = \int_{\Omega} (\partial_i v) u \, dA + \int_{\partial\Omega} u v n_i \, dl,$$

where ∂_i is the i th partial derivative and n_i the i th component of the normal to the boundary. For the cross term we can write either

$$\int_{\Omega} \mathcal{F}' \mathcal{G}' \, dA = \int_{\partial\Omega} \mathcal{G}' \mathcal{F} n_1 \, dl \quad \text{or} \quad \int_{\Omega} \mathcal{F}' \mathcal{G}' \, dA = \int_{\partial\Omega} \mathcal{F}' \mathcal{G} n_2 \, dl.$$

For their sum we have

$$2 \int_{\Omega} \mathcal{F}' \mathcal{G}' \, dA = \int_{\partial\Omega} \mathcal{F}' \mathcal{G} n_2 + \mathcal{G}' \mathcal{F} n_1 \, dl,$$

and since $\mathbf{n} = \pm(\mathcal{F}', \mathcal{G}')$, and $\mathcal{F} + \mathcal{G} = 0$ at the boundary, the integral equals zero. The energy is now simply

$$T\Psi = \frac{(1 + \lambda^2)}{2\lambda} \int_{\Omega} (\mathcal{F}'(\xi))^2 + (\mathcal{G}'(\eta))^2 \, d\xi \, d\eta. \quad (23)$$

Note that the energy is zero if the solution Ψ is constant. These solutions are excluded by (4), which also makes sure that the stream function cannot be scaled to get arbitrarily small energy. We also incorporate the corner constraint (3). We formulate a minimisation problem:

$$\begin{aligned} \min \quad & T\Psi, \quad \text{in } \Omega, \quad \text{and} \\ \min \quad & \Psi, \quad \text{at } \partial\Omega, \\ \text{s.t.} \quad & \\ & \Psi(\xi, \eta) = \mathcal{F}(\xi) + \mathcal{G}(\eta) \quad \text{in } \Omega \\ & \langle \mathcal{F}, 1 \rangle_{L_2([\xi^-, \xi^+])} = 0, \\ & \|\Psi\|_{L_2(\Omega)} = 1, \\ & \mathcal{F}' = \mathcal{G}' = 0 \quad \text{at a corner point.} \end{aligned}$$

As we saw in Section 5 this will lead to

$$\begin{aligned} \min \quad & \|Lx\|_2^2 \quad \text{energy minimisation,} \\ \min \quad & \|Ax\|_2^2 \quad \text{function minimisation at the boundary,} \\ \text{s.t.} \quad & Cx = 0 \quad \text{linear 'unique representation' constraint,} \\ & \text{and } \|x\|_2 = 1 \quad \text{normalisation,} \\ & \text{and } Px = 0 \quad \text{linear 'corner point' constraint.} \end{aligned} \quad (24)$$

The calculation of the matrix L is given in [24]. By analysing the singular value decomposition of a matrix corresponding to an ill-posed problem one can conclude that solutions corresponding to small singular values have many sign changes, and therefore possess high-frequency components (see [10] and the corresponding manual, which is also a good introduction to regularisation techniques). Some smoothing will be required. The smoothing is in the energy matrix L which contains discretisations of derivatives. If we have sufficient minimisation of the energy, the solution will be smooth. However, $\|Ax\|_2^2$ and $\|Lx\|_2^2$ cannot be minimised simultaneously. Also, by ill-posedness, there might exist a matrix close to A with much better properties in the sense that the energy $\|Lx\|_2^2$ is much lower while the residual $\|Ax\|_2^2$ is only slightly higher. Taking these considerations into account we propose to balance the residual $\|Ax\|_2^2$ and energy $\|Lx\|_2^2$ using a parameter τ . The regularised solution x_τ is then defined as

$$x_\tau = \arg \min(\|Ax\|_2^2 + \tau^2 \|Lx\|_2^2) \quad \text{with } \|x\|_2^2 = 1, \quad Cx = Px = 0. \quad (25)$$

The L-curve approach requires the solution of (25) for a range of parameters τ . In practice we choose for τ the numbers from the sequence $10^k, 10^{k-1}, \dots, 0$ with k approximately -10 or -15 (depending on the resolution of the grid). For lack of a better method of incorporating the $Px = Cx = 0$ constraint we add them to the matrix A with a large multiplicative factor a ,

$$x_\tau = \arg \min \left(\left\| \begin{pmatrix} A \\ aP \\ aC \end{pmatrix} x \right\|_2^2 + \tau^2 \|Lx\|_2^2 \right). \quad (26)$$

In this way the constraints will certainly be satisfied or the residual will become very large. Note that the value of a depends on the magnitude of the elements of A , which makes it dependent on the grid resolution. Solving the minimisation problem (26) is then equivalent to finding the normalised singular vector corresponding to the smallest singular value of

$$B = \begin{pmatrix} A \\ aP \\ aC \\ \tau L \end{pmatrix}. \quad (27)$$

6.1. The L-Curve

The choice of the value of the parameter τ is made using the *L-curve*. This is a continuous curve parametrised by τ and given by $\{(x(\tau), y(\tau)) | x(\tau) = \|Ax_\tau\|_2^2, y(\tau) = \|Lx_\tau\|_2^2\}$, where x_τ solves (26). The typical shape is that of an 'L', as shown in Fig. 4. This is an idealisation, in reality the L-curve will often be much less pronounced. The ideal

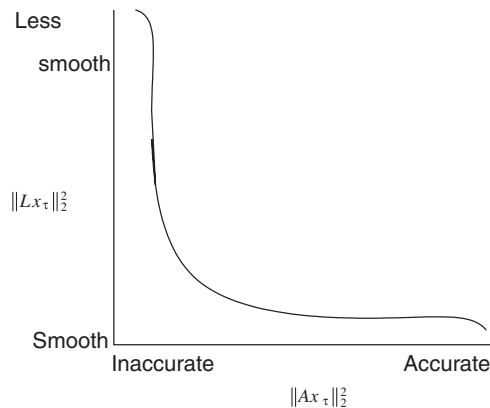


Fig. 4. This figure shows a schematic L-curve parametrised by τ . On the horizontal axis is the residual, representing the accuracy of the solution. On the vertical axis is the energy, representing the smoothness of the solution.

value of τ is found in the ‘elbow’ of the curve. Increasing τ would lead to a larger residual with only a slightly smaller energy (and thus a little gain in smoothness). Decreasing τ would lead to a slightly smaller residual, while the energy becomes very large, and we have less smoothness. In our experiments (see Section 8) we often find L-curves which have a less sharp corner, the ideal value of τ is not directly clear. Also, it might happen that the L-curve is not ‘L’ shaped at all. In these cases the curve can still be interpreted, and give us information on the solution. For example, some curves start horizontal, only to go down for very large values of τ . A mostly flat curve tells us that the value of the regularisation parameter is insubstantial, the solution has the same energy for every reasonable τ . The specific geometry under consideration did not give us an ill-posed problem. This happens often in the resonant case.

7. Energy minimisation

In this section we will show that, in the presence of attractors, the energy diverges. Nevertheless, we argue that the total energy is still a useful quantity. We can give qualitative descriptions of the solutions that minimise the energy. These solutions will turn out to be continuous and differentiable, which shows that minimising the energy is a useful idea. First, we clarify the procedure by a one-dimensional model.

7.1. A one-dimensional example

As we will show later, there are features of the minimal-energy solution to the hyperbolic equation that can be predicted. The generic two-dimensional case is rather complicated, for this reason this section will give a two-dimensional example that reduces to a one-dimensional model. The model will however exhibit the features that are also found in the full two-dimensional case. Some steps may seem rather trivial, yet they closely follow the more complicated steps in the two-dimensional case, treated in Appendix C of [24].

Consider the wave equation $\Psi_{xx} - \lambda^2 \Psi_{zz} = 0$ on a domain Ω with $\Psi = 0$ at the boundary $\partial\Omega$. We also set the normalisation constraint $\langle \Psi, \Psi \rangle = 1$. The Poincaré equation is separable in any rectangle, for illustration we use the square domain $\Omega = [0, 1]^2$. In this case, the Poincaré equation with the conditions stated above can be reduced to

$$\frac{f_{xx}(x)}{f(x)} = \lambda^2 \frac{g_{zz}(z)}{g(z)} \quad \text{in } \Omega, \quad (28)$$

$$g(0) = g(1) = f(0) = f(1) = 0, \quad (29)$$

$$\int_0^1 f^2(x) dx \int_0^1 g^2(z) dz = 1. \quad (30)$$

We would like to find the solution that minimises the energy

$$\begin{aligned} E &= \int_{\Omega} \nabla \Psi \cdot \nabla \Psi dx dy \\ &= \int_0^1 (f_x)^2 dx \int_0^1 g^2 dy + \int_0^1 f^2 dx \int_0^1 (g_y)^2 dy. \end{aligned}$$

Use the normalisation constraint (30) to see that minimising E is equivalent to minimising

$$\frac{\int_0^1 (f_x)^2 dx}{\int_0^1 f^2 dx} + \frac{\int_0^1 (g_y)^2 dy}{\int_0^1 g^2 dy}.$$

Now, this is really two separate minimisation problems, one for $f(x)$ and one for $g(y)$. We will concentrate on the function f and put $\delta = \int_0^1 (g_y)^2 dy$. Since we can vary x and y independently we can suppose the right-hand side of (28)

equals γ^2 . The simplified model now reads

$$\begin{aligned} f_{xx}(x) &= \gamma^2 f(x), \\ \|f(x)\|_{L_2(\Omega)}^2 &= \delta, \\ f(0) &= f(1) = 0. \end{aligned} \tag{31}$$

We need to find the solution that minimises

$$\tilde{E} = \delta \int_0^1 (f_x)^2 dx.$$

The differential equation plus boundary conditions (31) is solved by

$$f(x) = \sum_{k=0}^{\infty} a_k \sin(\gamma_k x),$$

with $\gamma_k = k\pi$. By (30), the $(a_i) \in \ell^2$ are subject to the constraint $\sum_{i=0}^{\infty} a_i^2 \neq 0$. Consequently the trivial solution is excluded, yet there are still infinitely many solutions. In order to pick one solution out of the infinite set of degenerate solutions we minimise the energy. Write the energy as

$$\tilde{E} = \delta \int_0^1 (f_x)^2 dx = \left\langle f, -\delta \frac{d^2}{dx^2} f \right\rangle_{L_2([0,1])}.$$

The Courant–Fischer theorem now gives us that the minimum of \tilde{E} is given by the eigenfunction belonging to the smallest eigenvalue of the operator $L = -\delta(d^2/dx^2)$, i.e., the function f corresponding to the smallest eigenvalue μ that solves

$$f_{xx}(x) = -\frac{\mu}{\delta} f(x). \tag{32}$$

This equation is of the same form as our differential (31), which is unfortunate from a didactical point of view: in the full two-dimensional model the situation is less trivial, and the Courant–Fisher theorem is a useful tool. The eigenvalue problem (32) has solution

$$f(x) = \alpha \cos\left(\sqrt{\frac{\mu}{\delta}} x\right) + \beta \sin\left(\sqrt{\frac{\mu}{\delta}} x\right).$$

From boundary conditions (29) we see that α must be zero and μ is quantised as $\mu_k = \delta k^2 \pi^2$ with $k = 1, 2, 3, \dots$, zero is excluded since the trivial solution is excluded. The eigenfunction f belonging to μ_k is then given by

$$f(x) = \beta_k \sin(k\pi x).$$

The constant β_k is found by solving $\int_0^1 f^2(x) dx = 1/d$ and we have a unique solution for given k . Considering solutions that belong to μ_1, μ_2, \dots we see that the energy orders the solutions by decreasing smoothness (in the sense of number of oscillations). The energy therefore has a regularising property.

Furthermore, we see that we find continuous and differentiable solutions ‘for free’, without explicitly enforcing these properties. These features carry over to the two-dimensional situation. However, when dealing with the two-dimensional hyperbolic equation in an arbitrary non-rectangular domain we have two additional issues

- The domain is arbitrary and we no longer have $\Psi(x, y) = f(x)g(y)$. However, in characteristic coordinates we do know $\Psi(\xi, \eta) = \mathcal{F}(\xi) + \mathcal{G}(\eta)$.
- Attractors will occur and the fractal behaviour of the solution introduces high energies.

Nonetheless, it will be shown that energy minimisation still leads to smooth regularised solutions. We will first give a one-dimensional analogue of the energy in the presence of an attractor and refer for the general situation to [24].

7.2. Behaviour near an attractor

Close to the attractor features of the solution are repeated at increasingly smaller scales. This makes the energy blow up, but we argue that we nonetheless still have a valid minimisation problem. We consider a one-dimensional model. For an arbitrary function $f \in H^1([0, 1])$ define:

$$\Psi(x) = \sum_{i=0}^{\infty} f_i(x),$$

$$f_i(x) = \begin{cases} f\left(\frac{x - \alpha_i}{\alpha_{i+1} - \alpha_i}\right) & \text{for } x \in [\alpha_i, \alpha_{i+1}], \\ 0 & \text{otherwise.} \end{cases}$$

Here $[\alpha_0, \alpha_1]$ models the fundamental interval, with $\alpha_2, \alpha_3, \dots$ iterates of its endpoints under the mapping T , as defined in Section 2. Usually the contents of an iterate of the fundamental interval is mirrored and scaled by the action of the mapping. We will not model the mirroring effect here, but suppose that an interval $[\alpha_i, \alpha_{i+1}]$ contains both the original function and its mirrored copy. The differences $\alpha_{i+1} - \alpha_i$ are steadily decreasing and there is an ‘attractor’ at $\lim_{i \rightarrow \infty} \alpha_i = \alpha$. The functions f_i are scaled and copies of f translated, this models the replication of the fundamental interval towards the attractor. The energy on the interval $[\alpha_0, \alpha_N]$ is modelled by

$$T([\alpha_0, \alpha_N]) = \int_{\alpha_0}^{\alpha_N} (\Psi_x)^2 dx$$

$$= \int_0^1 f'(x)^2 dx \sum_{i=0}^N \frac{1}{\alpha_{i+1} - \alpha_i}.$$

Here, we used the disjoint supports of f_i in order to write the square of sums as a sum of squares. There are two ways in which the energy may become infinite. Firstly, the derivative of the function f itself may not be square integrable, but such functions do not minimise the energy. Secondly, the sum in the second factor diverges, but for every finite value of the number of steps towards the attractor N , the expression is finite. Moreover, minimisation of the energy yields *the same* function f independently of N . Therefore, this function f may also be considered a solution for $N \rightarrow \infty$. Thus, even though we have a diverging total energy it is still useful for minimisation purposes.

From a numerical point of view, we propose that discrete solutions on a grid can be compared to the continuous solution on $[\alpha_0, \alpha_N]$ for finite N . Equivalently, one can say we always keep a finite distance to the attractor. The fine scale structure which causes the energy to diverge close to the attractor at limit point α is simply replaced with one basis function in the discretisation. Since solutions are independent of N our numerical solutions will correspond to solutions of the continuous system. In this way it makes sense to minimise the energy on a fixed grid, however, the energy of discrete solutions will increase with increasing grid resolution in the presence of attractors.

7.3. The two-dimensional case

In the two-dimensional case the calculation is more tedious, we will only state the results of the calculation here and refer for the technical details to Appendix C of [24]. Before giving the solution we introduce some notation. The α_i and β_i variables give ξ and η coordinates of the iterates of the fundamental interval. The index i counts the number of iterates towards the attractor and $k \leq K$ counts the fundamental intervals. The index $l \leq L$ takes care of the possibility that iterates of endpoints of a fundamental interval may approach a limit cycle with period L . For fixed k and l , the sequence $\alpha_i^{l,k}$ approaches one fixed point of the attractor from one side, $\cap_i(\alpha_i^{l,k}, \alpha_{i+1}^{l,k}) = \emptyset$ and $\cup_i[\alpha_i^{l,k}, \alpha_{i+1}^{l,k}]$ has no holes. The general solution can be described using wavenumber-like numbers (m, n) and arbitrary constants $A_n^{k,l}$ and $B_m^{k,l}$. It is given by

$$\Psi_{m,n}(\xi, \eta) = \mathcal{F}_n(\xi) + \mathcal{G}_m(\eta),$$

with

$$\begin{aligned}\mathcal{F}_n &= \sum_{i=0}^N \sum_{k=0}^K \sum_{l=0}^J A_n^{l,k} I_{\xi \in [\alpha_i^{l,k}, \alpha_{i+1}^{l,k}]} \cos \left(n\pi \frac{\xi - \alpha_i^{l,k}}{\Delta \alpha_i^{l,k}} \right), \\ \mathcal{G}_m &= - \sum_{i=0}^M \sum_{k=0}^K \sum_{l=0}^J B_m^{l,k} I_{\eta \in [\beta_i^{l,k}, \beta_{i+1}^{l,k}]} \cos \left(m\pi \frac{\eta - \beta_i^{l,k}}{\Delta \beta_i^{l,k}} \right),\end{aligned}\quad (33)$$

where I denotes the indicator function. The above formulas describe a solution which is built from arbitrary Fourier expansions on the fundamental intervals, which are reproduced in smaller scales towards the attractor. Note that $A_n^{k,l}$ cannot all be zero. The numbers N, M determine the size of neighbourhoods of the attractors that we leave out of the domain. For $N, M \rightarrow \infty$ we face a diverging energy, the above solutions are not valid. However, for every large *finite* values N, M we do have the solutions given above. With increasing N, M the solution does not change, except it's domain of definition grows, towards the attractor. For more details we refer to [24].

Some words on the space in which the functions \mathcal{F} and \mathcal{G} live are in order. In Section 2 we proposed $\mathcal{F} \in L_2([\xi^-, \xi^+])$ and $\mathcal{G} \in L_2([\eta^-, \eta^+])$, giving also $\Psi \in L_2(\Omega)$. The solution is discontinuous at the attractor, this follows easily from the definition of continuity and the fact that attractors are limit cycles of the transformation F . Theorem 4 established that solutions exist, and thus that the discontinuity is square integrable. In this section we need to calculate an energy, the suitable space would seem to be $H^1(\Omega)$. However, in this space the energy diverges when we insist on solving on the entire domain Ω (i.e., $N, M \rightarrow \infty$). The proper space seems to be $H^1(\Omega \setminus (\mathcal{A} \cup \mathcal{B}))$, where \mathcal{A} consists of a neighbourhood of the attractors. The set \mathcal{A} is in our approach left out of the domain by using finite N and M as shown above. It could also be the case that $\mathcal{F}(\xi)$ or $\mathcal{G}(\eta)$ are discontinuous. The set \mathcal{B} consists of horizontal and vertical lines corresponding to these discontinuities. This is only possible if $N, M \rightarrow \infty$ in (33), and we do not need to worry about excluding \mathcal{B} from the domain since we only consider finite values.

It is important to realise that \mathcal{A} and \mathcal{B} are completely different in nature. Attractors are a feature of the problem, but we know how to deal with this. Discontinuities in \mathcal{F} and \mathcal{G} away from the attractor will not occur because the minimising functions (33) are automatically continuous.

7.4. Minimal energy of the discretised system

Consider the behaviour of the solution to the discrete system. When piecewise smooth basis functions are used, then across the attractor there will be a piecewise smooth approximation of the self similar structure. We suggest that this discrete system corresponds, to good approximation, to solving the continuous system where we only minimise the energy on the domain given by a finite number of iterates of the fundamental interval. This corresponds to taking N and M to be finite numbers. We have now taken away the contributions to the energy that caused it to diverge, and have in effect removed a small strip around the attractors. Still we know that the minimal-energy solution we find is independent of the values of M and N , if $M \gg 1$ and $N \gg 1$. If the grid has sufficient resolution, then we will find a meaningful solution to the Poincaré equation. The energy will increase with decreasing grid size, yet for a fixed grid we can safely minimise the energy. If differences between calculated solutions on grids of increasingly higher resolution become smaller, then we can have confidence that we are converging towards a minimal-energy solution of the continuous problem.

8. Results

This section will give some solutions of the Poincaré equation using the discretisation from Section 5 and the regularisation procedure outlined in Section 6. First, we consider the case where the stream function is zero on the entire boundary, i.e., the unforced setting. Next we apply a non-zero boundary condition for the stream function and find that the scenario sketched in Section 3 indeed occurs.

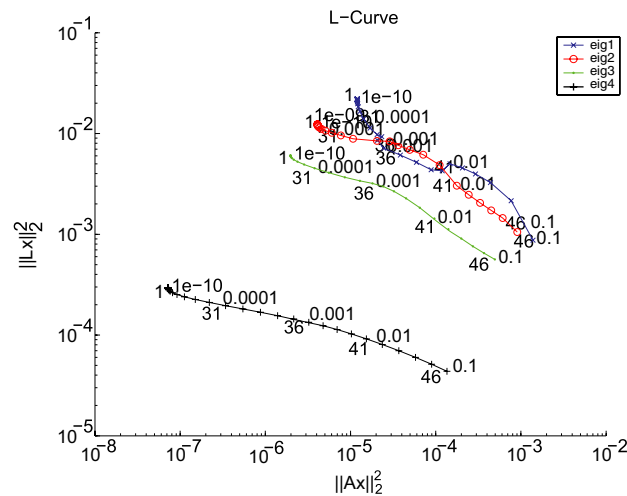


Fig. 5. Four L-curves for the trapezoid. The ‘elbows’ of the curves do not look very pronounced when plotted in one picture. One would have expected a longer vertical ‘leg’, instead we find an accumulation point. A number of points on the curves have been calculated and labelled with the value of the regularisation parameter τ and an index.

8.1. Free oscillations

We will discuss the trapezoid, a rectangle with one sloping side as shown in (x, z) coordinates in Fig. 10. This geometry is interesting since it is a simple geometry in which attractors exist. The complex distribution of attractor regimes separated by modal solutions (a manifestation of the ill-posed nature of the problem) is extensively described in [18]. For $\lambda=0.75$ we are in the attractor case, indeed Fig. 10 (as later discussed) clearly shows features that reproduce in increasingly smaller scales towards a limit cycle. The behaviour of the dynamical system induced by F was investigated by a raytracing technique, which enables us to find the values of λ where the various attractors exist.

We take 300 grid points in both the ξ and η directions and plot L-curves of the first four solutions. This means that we plot the L-curves of the first few solutions that minimise (26) best. This simply amounts to calculating the four smallest singular values of the matrix (27). The curves are approximated by taking a sequence of values for τ , we choose $\{10^j | j = -10, -9.8, -9.6, \dots, -0.8, -1\}$. The L-curves are given in Figs. 5 and 6 gives enlargements of the parts of the curves where we pick the regularisation parameter.

We will discuss the solutions for the different L-curves and point to noteworthy features of the solutions.

8.1.1. The ‘plateau’ solution

From the bottom curve of Fig. 5, the top-left curve of Fig. 6, representing the minimising solution, we choose a value of $\tau \approx 10^{-5}$. We plot the functions \mathcal{F} and \mathcal{G} and their sum in Fig. 7. We observe a solution with a large step at the attractor, featuring small oscillations. Two questions come to mind, firstly if such a solution is allowed. Secondly, if the small oscillations are a numerical artifact (reminiscent of the Gibbs phenomenon) or a real feature of the solutions. The latter question is answered by observing the structure of the oscillations: they are mapped towards the attractor like the oscillations in Figs. 8 and 11. We conclude that the structure is ‘mixed in’ from another solution, the regularisation is not perfect. The question of the validity is more complicated. The solution is of the form (33), with large coefficients for the constant term and very small coefficients for other terms. In the domain outside of the attractor the values \mathcal{F} plus \mathcal{G} cancel, while inside of the attractor they add up to a non-zero value. Physically, this represents a solution where (almost) all energy is located at the attractor. A problem is that the analytical analogue of this solution is not from the space $H^1(\Omega)$, the energy would be infinite. As argued in Section 7 we ought to compare discrete solutions in Ω to analytical solutions in Ω minus a neighbourhood of the attractor, and in this sense the solution is valid. In [23] it is shown that this type of solution is also found upon inclusion of a viscous (dissipative) term in the Poincaré equation. This is an indication that this might also physically be a relevant solution. The experiment described in [18] is a case in point, the existence of a wave attractor was established, but no evidence of fractal structure was found.

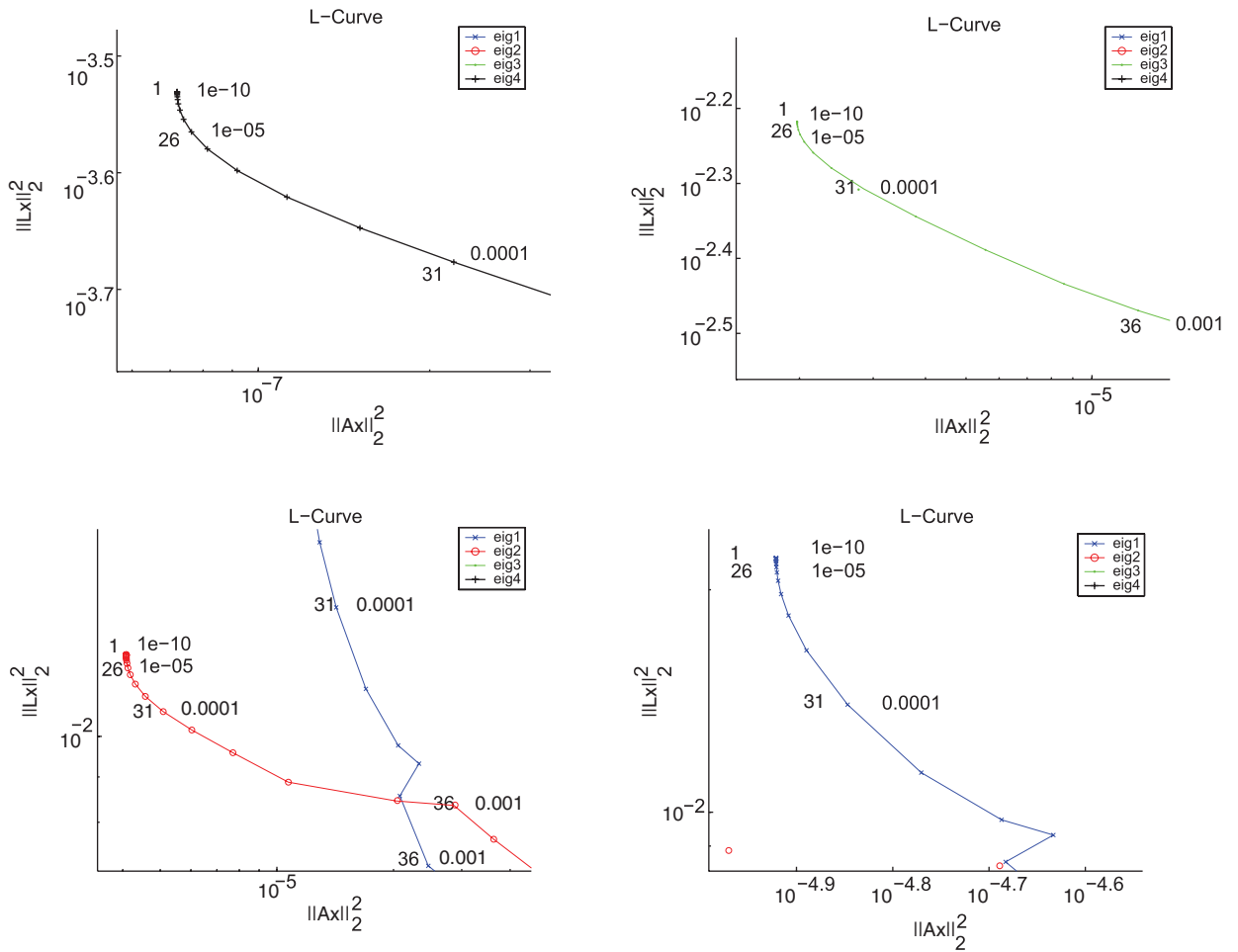


Fig. 6. These four panels show the L-curves corresponding to the best four solutions. The indices of the regularisation parameters are (in clockwise order): 26, 31, 31, 31, corresponding to $\tau \approx 10^{-5}, 10^{-4}, 10^{-4}, 10^{-4}$.

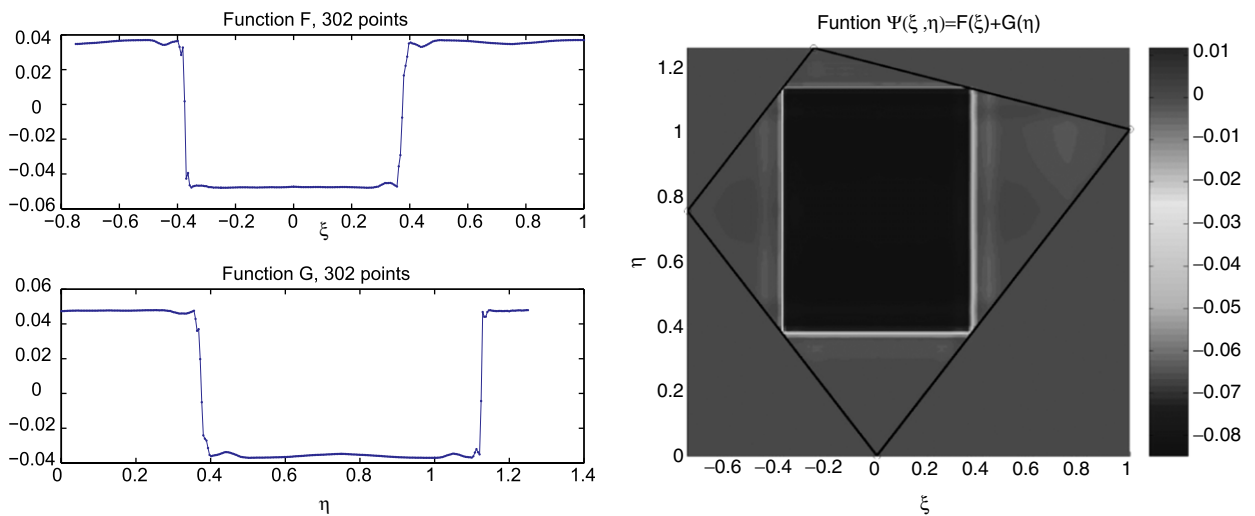


Fig. 7. This figure shows the functions \mathcal{F} and \mathcal{G} and a plot of $\Psi = \mathcal{F} + \mathcal{G}$, in the (ξ, η) coordinate frame. Darker colors indicate higher values, grey is at the zero level.

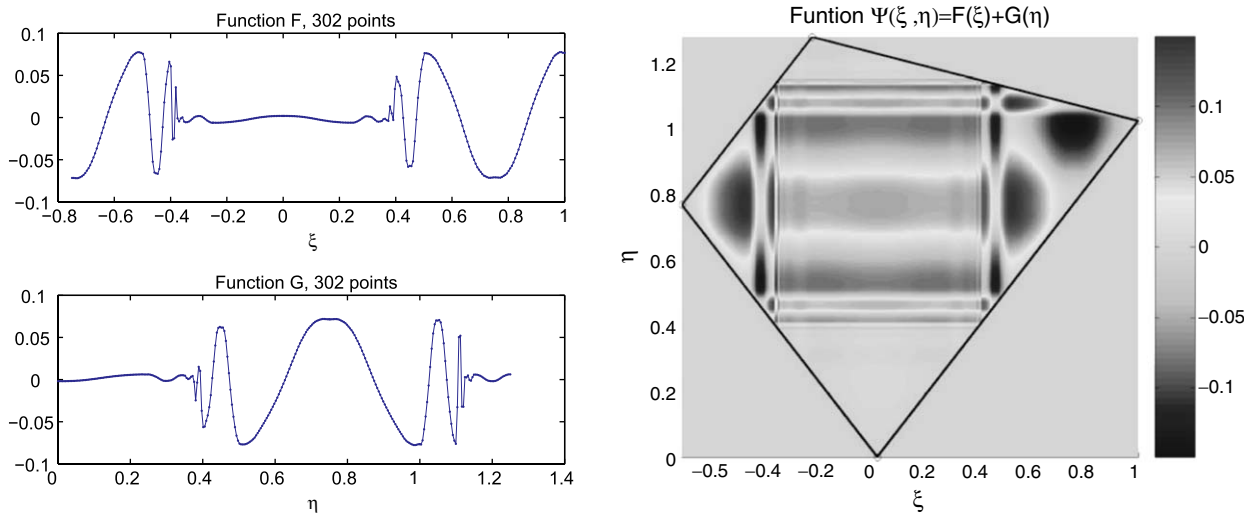


Fig. 8. This figure shows the functions \mathcal{F} and \mathcal{G} and a plot of $\Psi = \mathcal{F} + \mathcal{G}$, in the (ξ, η) coordinate frame. One fundamental interval shows a smooth function (half the period of a cosine), the other one is zero.

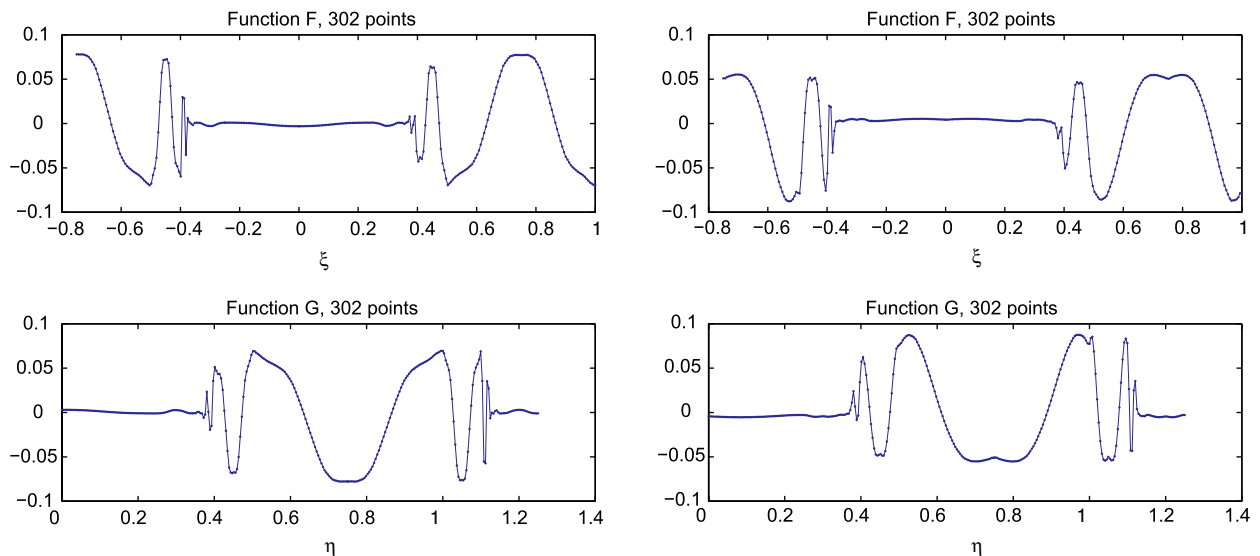


Fig. 9. These graphs show what happens if one regularises improperly. In the left panel a value of $\tau \approx 3 \times 10^{-3}$ was chosen, yielding a smooth but inaccurate solution. The extrema are too flat. The right panel corresponds to $\tau \approx 10^{-5}$, the residual is smaller but the energy is higher. We see cusps appearing, giving a discontinuity in the derivative.

8.1.2. Wavenumber 1 solution

The next-best solution we find is shown in Fig. 8, where we plot the solution corresponding to $\tau \approx 10^{-4}$ on the second curve (Fig. 6). It clearly consists of one empty fundamental interval plus one fundamental interval with half a cosine. Referring to (33) we call this a wavenumber 1 solution for this fundamental interval. We will now demonstrate the effect of over- and under-regularisation by picking inappropriate values of τ . The result of this practice is shown in Fig. 9. Other experiments have revealed even more sensitive dependence of the solution on the choice of the regularisation parameter. When the grid has sufficient resolution one may also observe oscillations that are introduced

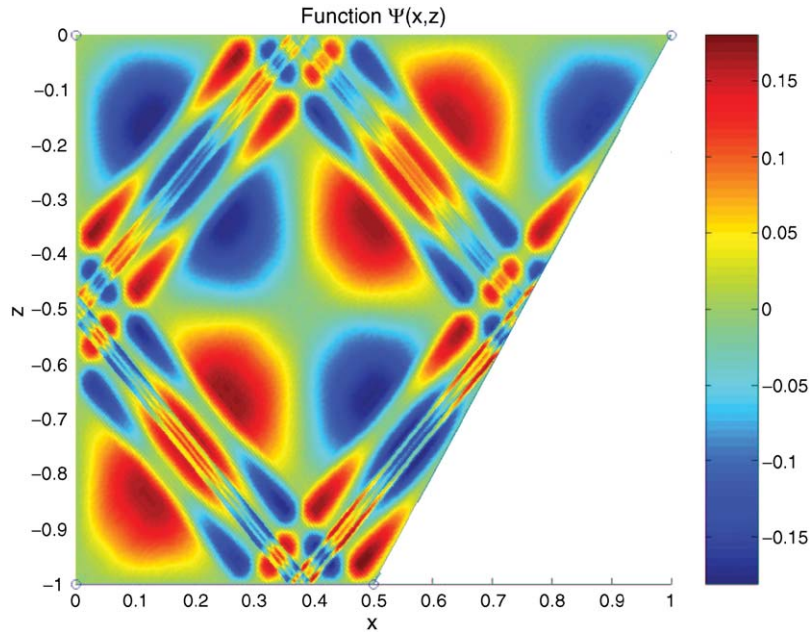


Fig. 10. This figure shows the addition of the smoothest solutions found on each of the fundamental intervals. The grid resolution is 300 grid points in the ξ and η directions. Shown is the stream function Ψ , in the (x, z) frame. Darker colors indicate higher values, grey is at the zero level. The attractor, and the fractal structure are nicely visible.

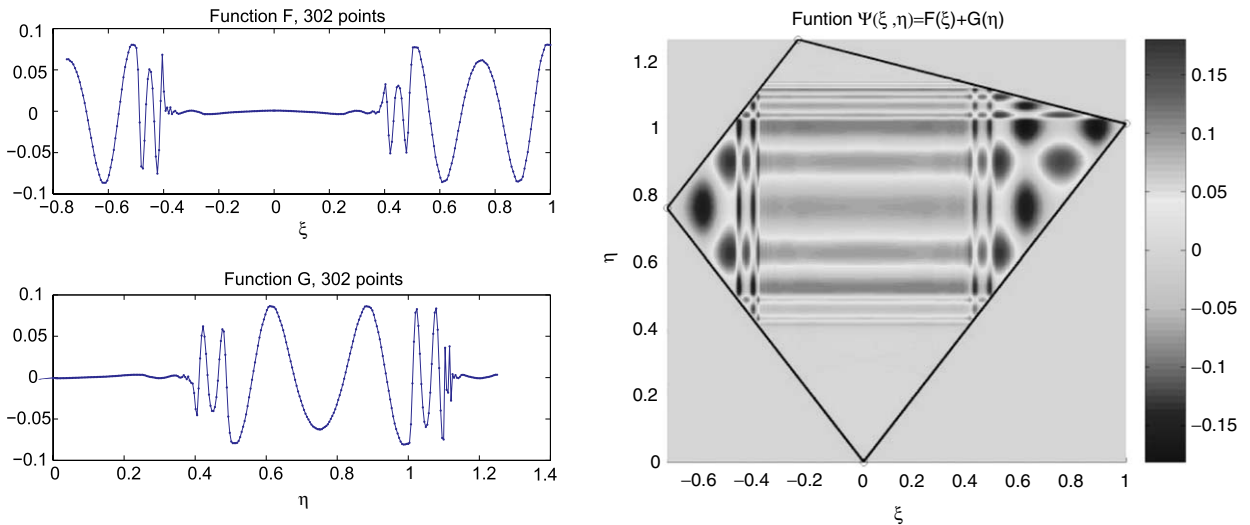


Fig. 11. In this figure, again one fundamental interval shows a smooth function, but now a cosine of higher periodicity. The other interval is (almost) zero.

or suppressed when changing the regularisation parameter. These observations are ultimately the justification for the use of a regularisation procedure.

8.1.3. Combined solutions

Fig. 5 shows the curves corresponding to the four best minimizing solutions. The top two curves cross each other twice, which might seem strange at first sight, but there is no reason why a under-regularised n th best solution cannot

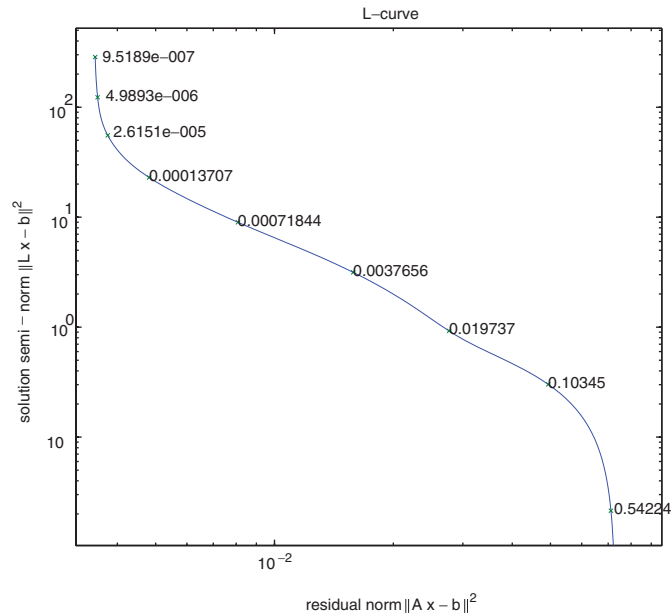


Fig. 12. The L-curve for the Poincaré problem with non-zero boundary condition on part of the boundary.

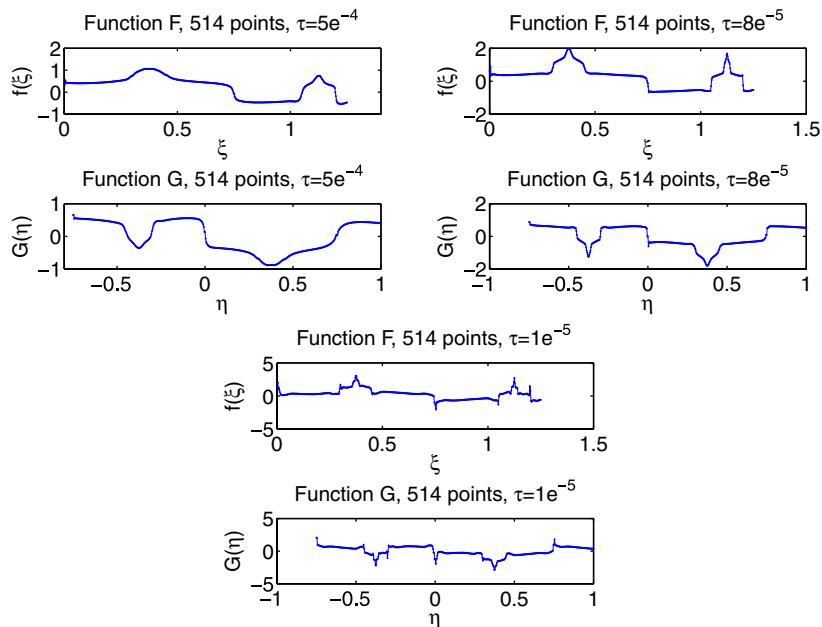


Fig. 13. These panels show the functions \mathcal{F} and \mathcal{G} for three different values of the regularisation parameter. To the left we see under-regulation, a value of $\tau = 5 \times 10^{-4}$ was used, which is below the optimum (see also Fig. 12). The graphs are clearly too smooth. A value of $\tau = 8 \times 10^{-5}$ seems close to optimal and yields the graphs in the right panel. Note how the functions \mathcal{F} and \mathcal{G} have jumps of ± 1 as predicted in the theory. Over-regulation is shown in the bottom panel, at $\tau = 1 \times 10^{-5}$, the functions suffer from too many sharp peaks.

have a higher energy than an over-regulated $(n + 1)$ th best solution. The top-most crossover of the curves is in fact an artefact of the way the L-curves are plotted. For each value of τ we store a column of coordinates $(\|Ax_\tau\|_2^2, \|Lx_\tau\|_2^2)$ in a matrix, we then plot the rows of this matrix. The right panel of Fig. 5 shows that the eigenvalue curves and L-curves switch at index 35.

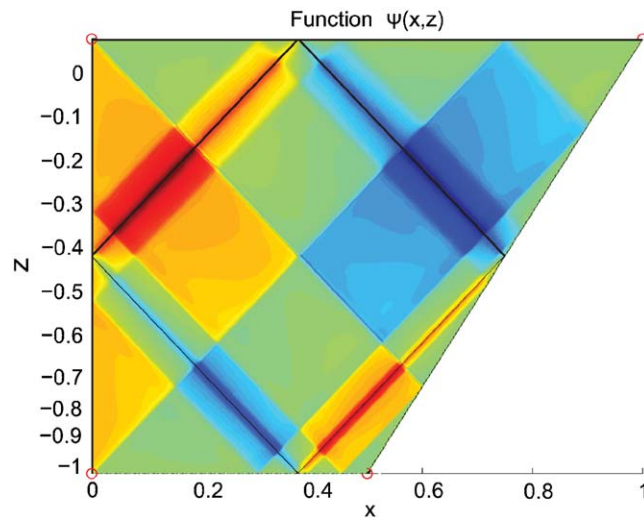


Fig. 14. A plot of the stream function for the forced problem. At the vertical left boundary $\Psi = 1$ is assigned, the remaining boundary has $\Psi = 0$. It is nicely visible how the characteristics ‘pick up’ values of ± 1 when visiting the side where $\Psi = 1$. Also, we clearly see focusing towards the attractor, where values of $|\Psi|$ get increasingly higher due to continued visits to the left boundary.

At eigenvalue curve three we find the solution where the empty fundamental interval has switched places with the wavenumber 1 interval. If we add both solutions (see Fig. 10), then we fill both fundamental intervals and recover the solution that was presented in [17].

8.1.4. Wavenumber 2 solution

The solution we find, at $\tau \approx 10^{-4}$, curve four of Fig. 6, is a solution where again one of the fundamental intervals is empty. The other one however is now filled with a complete period of a cosine (Fig. 11). We remark that the order in which we find solutions is not a priori clear. In this case we first found both the ‘wavenumber 1’ solutions, we could also have found a ‘wavenumber 2’ solution first. The exact order is dependent on the energy of the solution, which is (among other things) determined by the sizes of the fundamental intervals.

We conclude that we find the solutions predicted by Eq. (33), where the \mathcal{F}_n and \mathcal{G}_m parts are indeed decoupled. The regularisation scheme is indispensable in obtaining meaningful solutions to the ill-posed Poincaré equation.

8.2. Boundary forcing

In this section, we apply a boundary condition $\Psi = 1$ to the left side of the trapezoid. We have to deal with Eq. (8), which leads to the regularisation problem (9). In contrast to the unforced case, with zero right-hand side b , this is a standard problem which may be tackled using existing methods. We draw the L-curve and calculate solutions using the *Regularization Tools*, described in [10]. In this section we focus on the piecewise constant solution, since we learned in the previous section that this is energetically the preferred solution.

We calculate a solution at $\lambda = 0.75$, which is in the regime where the square attractor lives. The number of grid points is 512 in both the ξ and η directions. The L-curve calculated for this problem is shown in Fig. 12. This L-curve has a clearer optimal point than the curves presented before, although it is still necessary to try a few values around the bend in the curve before deciding on a ‘best solution’. Results for the functions \mathcal{F} and \mathcal{G} are shown in Fig. 13, where the effects of over-regularisation and under-regularisation are nicely visible.

Finally, we present a figure of the complete solution $\Psi(\xi, \eta) = \mathcal{F}(\xi) + \mathcal{G}(\eta)$ in Fig. 14. The theoretically predicted behaviour (Section 3) is clearly observed.

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